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High-order discontinuous Galerkin method for entry simulations into dusty atmospheric environments

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Objective

Summary of research results from NASA-supported ECF award NNX15AU58G on "Advanced Physical Models and Numerical Algorithms to Enable High-Fidelity Aerothermodynamic Simulations of Planetary Entry Vehicles on Emerging Distributed Heterogeneous Computing Architectures"

Overall research goals

 Develop innovative physical models and advanced numerical methods for reliably predicting aerothermodynamic flows that are relevant to hypersonic and atmospheric entry vehicles in dusty flow environments

Objective

- Develop high-order DG solver for high-speed disperse multiphase flows
 - Shock capturing method, with a focus on predicting surface heating in hypersonic flows
 - Lagrangian particle method suited for arbitrary curved, high-aspectratio elements
- Assess DG method for hypersonic, particle-laden flow applications
 - Quantitative comparisons with state-of-the-art finite volume solvers (FUN3D, LAURA)
- Apply developed DG framework to hypersonic dusty flows simulations
 - > Parametric sensitivity study
 - Hypersonic dusty flow over the ExoMars Schiaparelli capsule

Acknowledgments

NASA

Michael Barnhardt, Grant Palmer, Khalil Bensassi, Peter Gnoffo, Ethiraj
 Venkatapathy, Amal Sahai, and entire EDL-team

DLR

Ali Gülhan, Dirk Allofs

Stanford

Yu Lv, Steven Brill, Narendra Singh, Alex Aiken

Outline

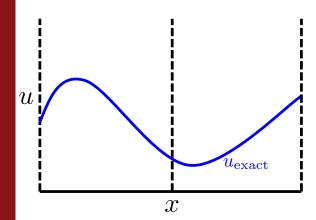
Development Background **Application** Hypersonic dusty flows over blunt bodies Introduction & Shock capturing Motivation scheme Experimental data and parametric Lagrangian DG mathematical sensitivities particle method background Mars atmospheric entry

What is DG?

Discontinuous Galerkin methods are a family of numerical methods combining features of finite-element and finite-volume schemes

- Piecewise polynomials used to approximate the solution
 - Arbitrary polynomial order p
 - In each element, solve for polynomial coefficients \(\tilde{u} \)

$$u_{\text{approx}}(t,x) = \sum_{i=1}^{p+1} \tilde{u}_i(t)\phi_i(x)$$



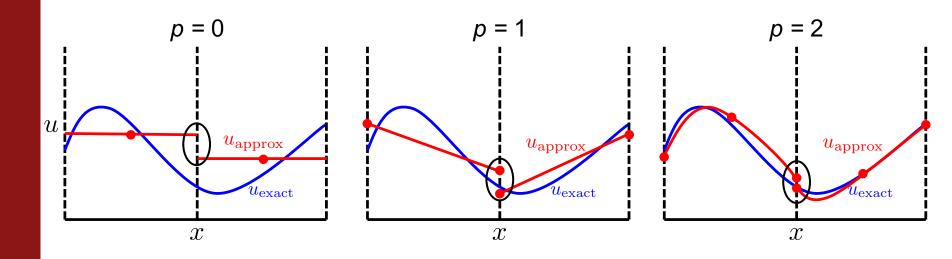
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Numerical fluxes needed to deal with discontinuities between elements



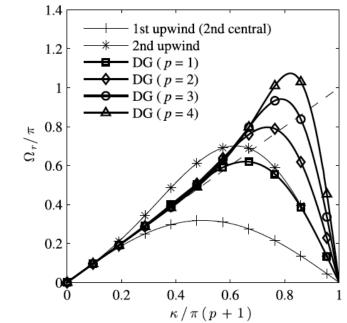
Why DG?

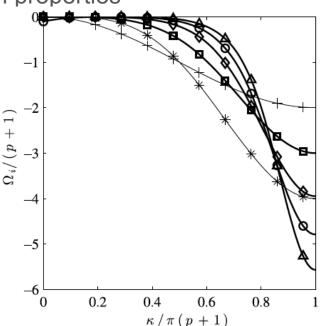
Current state of simulations of multi-physics (turbulent, multiphase, chemically reacting, etc.) flows on complex geometries

- Finite-difference, finite-volume methods
- Low-order (1st or 2nd order)

Potential for high-order DG methods

Desirable dissipation and dispersion properties





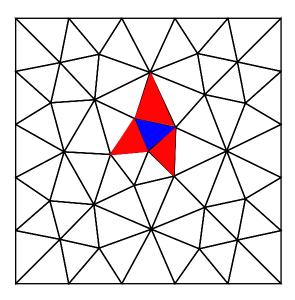
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Potential for high-order DG methods

- Desirable dissipation and dispersion properties
- Suited for HPC (compactness, FLOPs : memory bandwidth)



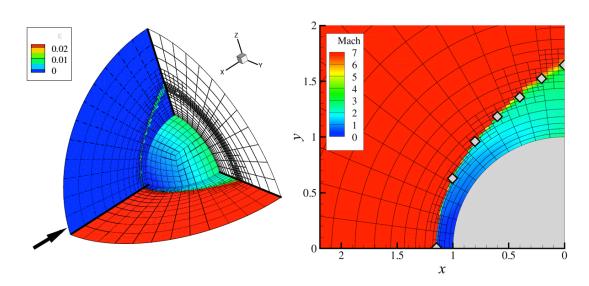
Why DG?

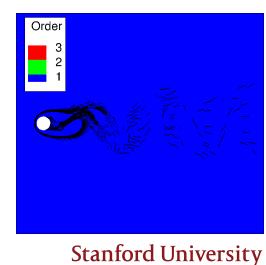
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Potential for high-order DG methods

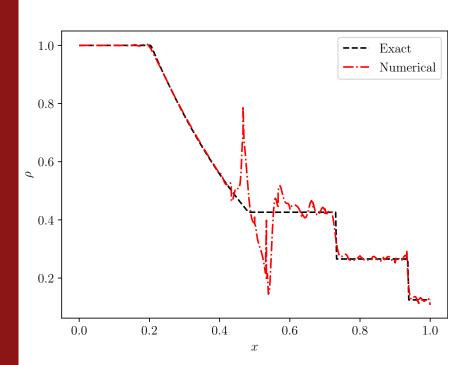
- Desirable dissipation and dispersion properties
- Suited for HPC (compactness, FLOPs : memory bandwidth)
- Local mesh-adaptation (h) and refinement in polynomial order (p)

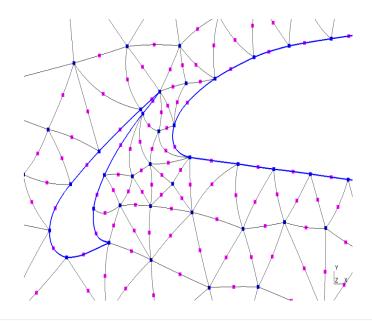


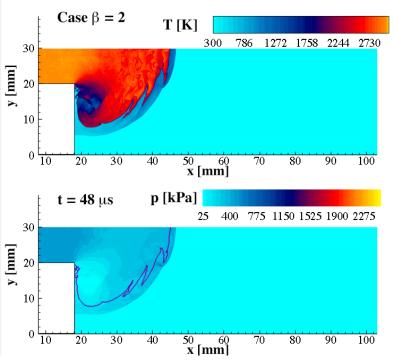


Challenges

- Efficient time stepping
- Efficient automatic hp-adaptation
- High-order curved meshes
- Numerical instabilities
- Extensions to complex physics







[1] Lv and Ihme, Proc Combust Inst, 2015

Outline

Development Background **Application** Hypersonic dusty flows over blunt bodies Introduction & Shock capturing Motivation scheme Experimental data and parametric DG mathematical Lagrangian sensitivities background particle method Mars atmospheric entry

Discontinuous Galerkin discretization

Governing equations

$$\partial_t oldsymbol{U} +
abla \cdot oldsymbol{F}_s =
abla \cdot oldsymbol{F}_v + oldsymbol{S}$$
 $\left[egin{array}{c}
ho oldsymbol{u} \end{array}
ight] \left[egin{array}{c}
ho oldsymbol{u} \end{array}
ight]$

$$egin{aligned} oldsymbol{U} & = egin{bmatrix}
ho oldsymbol{u} \
ho oldsymbol{u} \
ho oldsymbol{E} \end{bmatrix}, & oldsymbol{F}_s & = egin{bmatrix}
ho oldsymbol{u} \
ho oldsymbol{u} \otimes oldsymbol{u} + P \mathbb{I} \ oldsymbol{u} \
ho E \end{bmatrix}, & oldsymbol{F}_v & = egin{bmatrix} 0 \ oldsymbol{ au} \ oldsymbol{u} \cdot oldsymbol{ au} - oldsymbol{q} \end{bmatrix} \end{aligned}$$

Partition computational domain

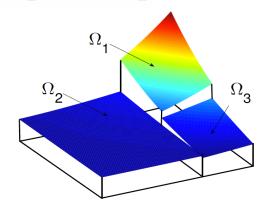
$$\Omega = \bigcup_{e=1}^{N_e} \Omega_e$$

Approximate solution with polynomials of order *p*

$$oldsymbol{U}pproxoldsymbol{U}_h=\oplus_{e=1}^{N_e}oldsymbol{U}_h^e$$

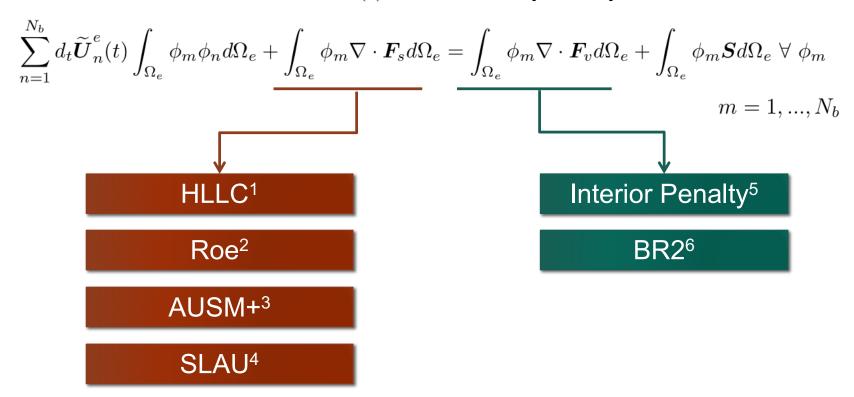
$$oldsymbol{U}_h^e(oldsymbol{x},t) = \sum_{n=1}^{N_b} \widetilde{oldsymbol{U}}_n^e(t) \phi_n(oldsymbol{x})$$

Multiply governing equations by ϕ_m and integrate



Discontinuous Galerkin discretization

Find the basis coefficients $\widetilde{m{U}}^e(t)$ that discretely satisfy



Evaluate integrals using Gaussian quadrature

[1] Toro, et al., SW, 1994;

[4] Shima and Kitamura, AIAA J., 2011;

[2] Roe, JCP, 1981; [3] Liou, JCP, 1996; [5] Arnold, et al. SIAM JNA, 1981;[6] Bassi & Rebay, JCP, 1997.

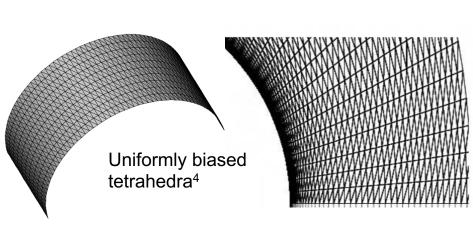
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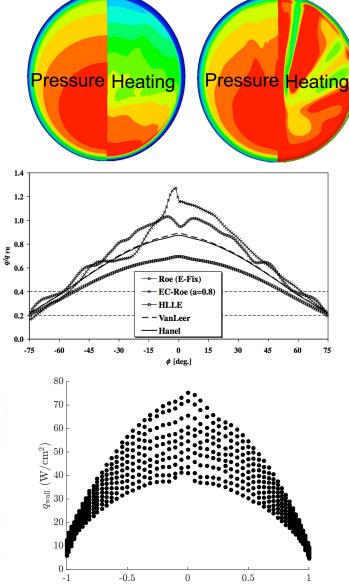
Outline

Development Background **Application** Hypersonic dusty flows over blunt bodies Shock capturing Introduction & Motivation scheme Experimental data and parametric Lagrangian **DG** mathematical sensitivities particle method background Mars atmospheric entry

Background: finite-volume

- Surface heating predictions extremely sensitive to:
 - Mesh-shock alignment¹
 - Inviscid flux function^{2,3}
 - Limiter, parameters, etc.^{2,3}
- Gnoffo computed hypersonic flow over a cylinder using uniformly biased tetrahedra⁴





Alignment

No alignment

- [1] Candler et al., AIAA Paper, 2009
- [2] Kitamura et al., AIAA Journal, 2010
- [3] Kitamura, AIAA Paper, 2013
- [4] Gnoffo and White, AIAA Paper, 2004

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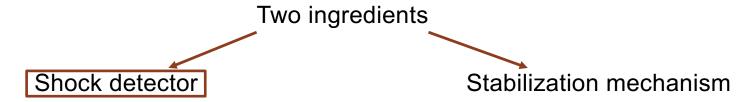
Background: DG

- Robust shock capturing is still an active area of research
- Limited work on using high-order DG to compute viscous hypersonic flows^{1,2,3,4}

Goals:

- Develop a simple and robust shock capturing method
- Assess sensitivities of DG heating predictions to mesh-shock alignment and choice of inviscid and viscous flux functions
- Perform quantitative comparisons with FUN3D and LAURA

Shock capturing method



$$S_e = \sqrt{\frac{\int_{\Omega_e} (\mathbf{s}/\bar{\mathbf{s}} - 1)^2 d\Omega}{|\Omega_e|}}$$

Intra-element variations

- Simple
- Compact
- Large near discontinuities, small in smooth regions

Identify troubled elements

Shock capturing method

Two ingredients

Shock detector

Stabilization mechanism

Artificial viscosity (AV) term $\nabla \cdot oldsymbol{F}_{ ext{AV}}$

$$oldsymbol{F}_{ ext{AV}} = oldsymbol{H}:
abla oldsymbol{U}$$

$$oldsymbol{H}_{ik} = \eta_{ ext{AV}} rac{h_{ik}}{ar{h}} \mathbb{I}$$

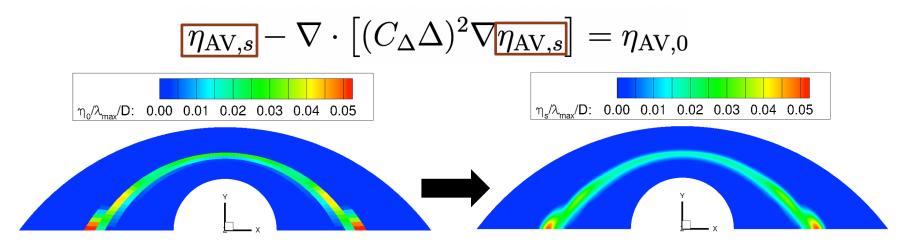
Add initially elementwise-constant AV to troubled elements

$$\eta_{\text{AV},0}^e = C_\eta \lambda_{\text{max}} \frac{\bar{h}}{\max(1,p)} \bar{S}_e$$

Shock capturing method



Smooth AV



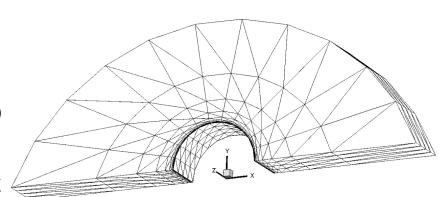
Reduce AV in boundary layers

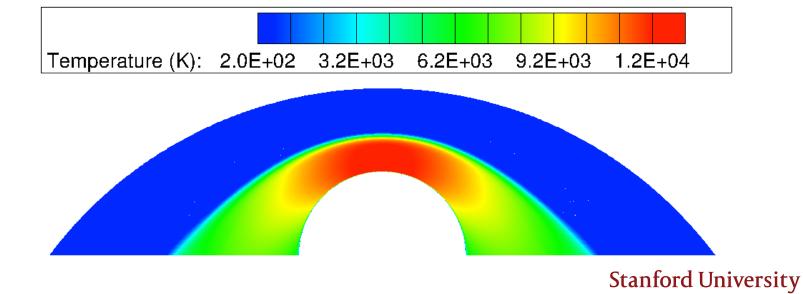
- Homogeneous Dirichlet BCs at no-slip walls
- Pass $\eta_{{
 m AV},s}$ through sinusoidal filter² to get $\eta_{{
 m AV}}$

Test case: Mach 17.6 flow over circular half-cylinder

Uniformly biased tetrahedral mesh

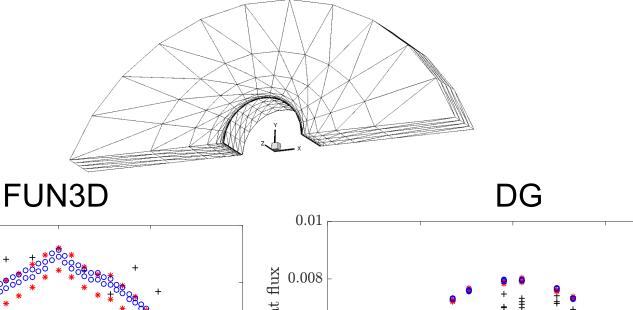
- DG: (*p*+1)-order accuracy, where *p* = 1, 2, 3
- Quantitative comparisons with FUN3D
 - 2nd-order accuracy
- Influences of inviscid and viscous flux functions on heating predictions

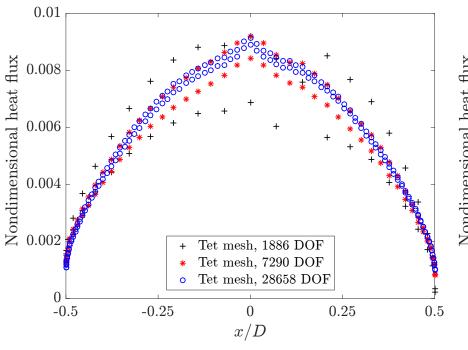


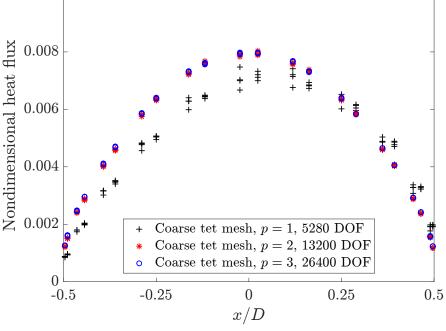


Ma *p*-order 17.6 1-3

FUN3D heat flux comparisons



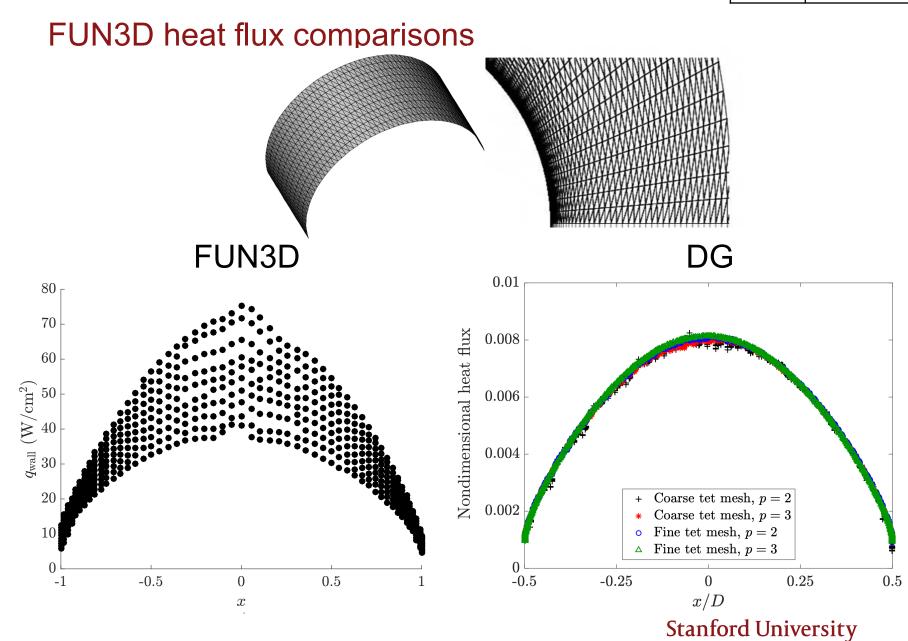




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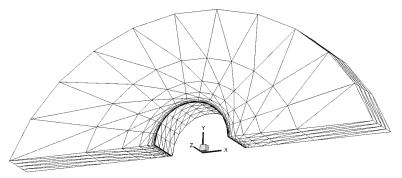
 Ma
 p-order

 17.6
 1-3

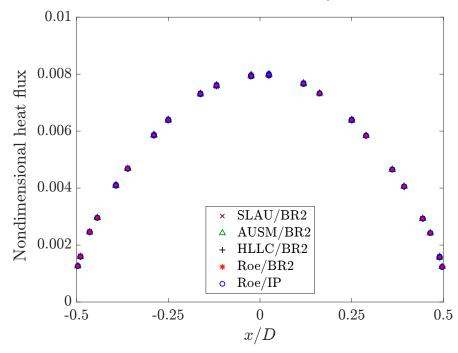


Ma	<i>p</i> -order
17.6	3

Sensitivity to inviscid and viscous flux functions

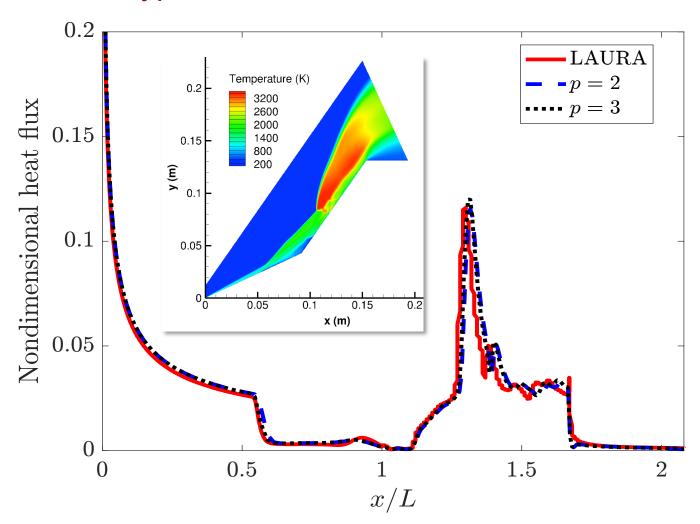


Coarse tet mesh, p = 3



Ma	<i>p</i> -order
9.59	2-3

Test case: hypersonic flow over double cone



→ DG solution reaches convergence with ~half # DOF of LAURA solution

Summary

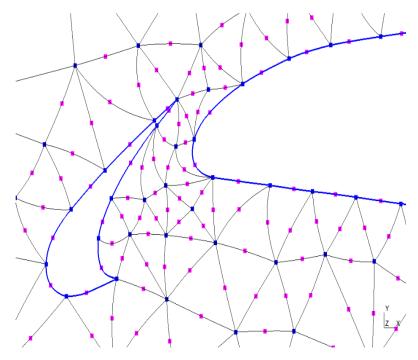
- Developed a simple and robust robust shock capturing method for DG
- Applied the proposed formulation to viscous hypersonic flows
- DG heating predictions are much less sensitive to mesh-shock alignment and choice of inviscid flux function than FV heating predictions
- Fewer degrees of freedom are required to achieve mesh convergence in DG solutions

Outline

Development Background **Application** Hypersonic dusty flows over blunt bodies Introduction & Shock capturing Motivation scheme Experimental data and parametric Lagrangian **DG** mathematical sensitivities particle method background Mars atmospheric entry

Background: particle-laden flows

- Several point-particle Euler-Lagrange formulations that rely on highorder numerical methods
 - Finite-difference^{1,2}
 - Spectral methods^{3,4}
- Limited development in the context of:
 - DG^{5,6,7}
 - Curved elements



[7] Zwick and Balachandar, *Int J High Perform Comput Appl*, 2020

^[1] Jacobs and Don, J Comp Phys, 2008.

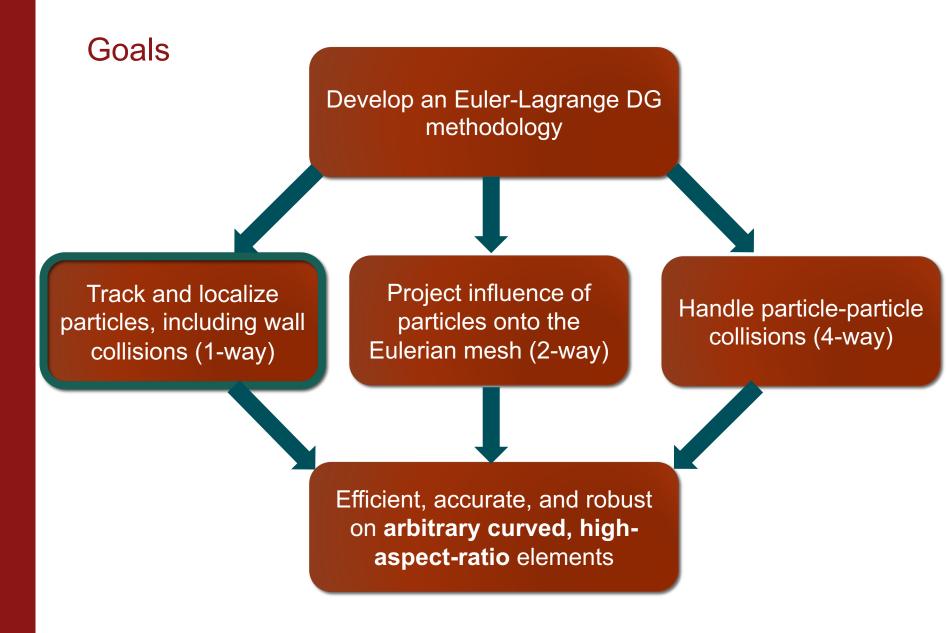
^[2] Jacobs et al., Theor Comp Fluid Dyn, 2012

^[3] Bagchi and Balachandar, J Fluid Mech, 2002

^[4] Akiki et al., J Comp Phys, 2017

^[5] Sengupta et al., Int J Multiph Flow, 2009

^[6] Huang et al, Comput Methods Appl Mech Eng, 2019



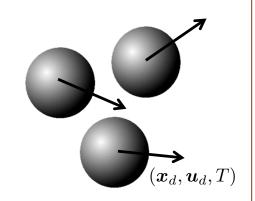
Lagrangian Particle Tracking

- Individual particles ("d") are tracked in the flow field as point sources
- Carrier gas ("c") described with Eulerian field

Position:
$$\frac{d{m x}_d}{dt} = {m u}_d$$

Momentum:
$$m_d \frac{d oldsymbol{u}_d}{dt} = oldsymbol{F} = oldsymbol{F}_{qs} + oldsymbol{F}_{ ext{other}}$$

Energy:
$$m_d c_d \frac{dT_d}{dt} = Q = Q_{qs} + Q_{\text{other}}$$



Quasi-steady contributions

Drag:
$$m{F}_{qs} = rac{1}{8}\pi D^2
ho_c (m{u}_c - m{u}_d) |m{u}_c - m{u}_d| C_D$$

Heating Rate: $Q_{qs} = \pi D \kappa_c (T_c - T_d) \mathrm{Nu}$

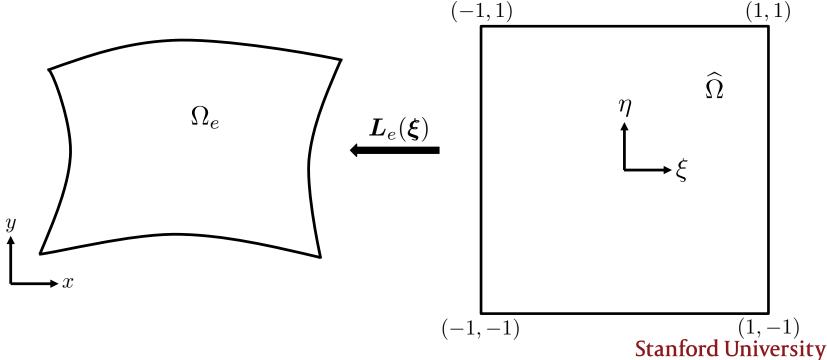
$$C_D = f(Re_d, Ma_d)$$
 $Nu = f(Re_d, Ma_d)$

Particle search-locate procedure

Algorithm by Allievi and Bermejo

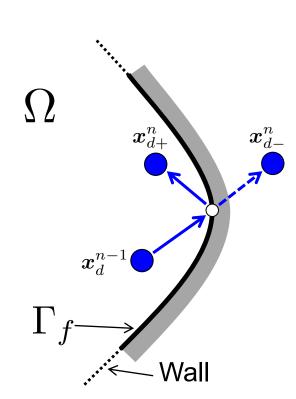
- Compatible with arbitrary curved elements
- Relies on the geometric mapping

$$\boldsymbol{L}_{e}(\boldsymbol{\xi}) \equiv \begin{bmatrix} x \\ y \end{bmatrix} = \sum_{i=1}^{N_{n}} \psi_{i}(\boldsymbol{\xi}) \begin{bmatrix} x_{i}^{e} \\ y_{i}^{e} \end{bmatrix}$$



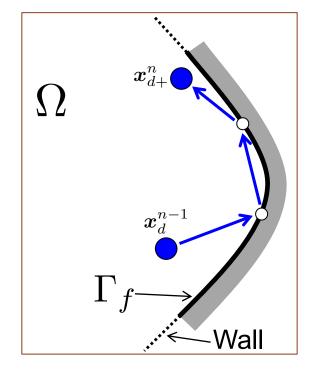
Particle-wall collisions

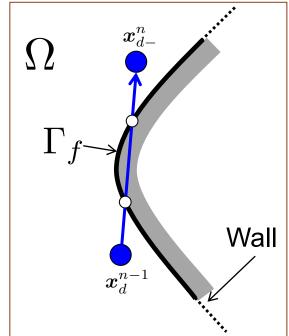
- Extended the search-locate procedure to account for hard-sphere particle-wall collisions
- Apply Newton search to compute intersection between particle trajectory and boundary
- Curved, high-aspect-ratio elements
 - Pathological cases¹



Particle-wall collisions

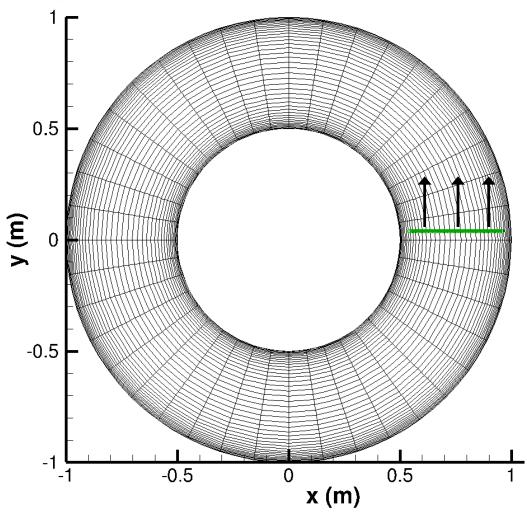
- Extended the search-locate procedure to account for hard-sphere particle-wall collisions
- Apply Newton search to compute intersection between particle trajectory and boundary
- Curved, high-aspect-ratio elements
 - Pathological cases¹
 - Finite size of particles²





Test case: particle advection in annular channel

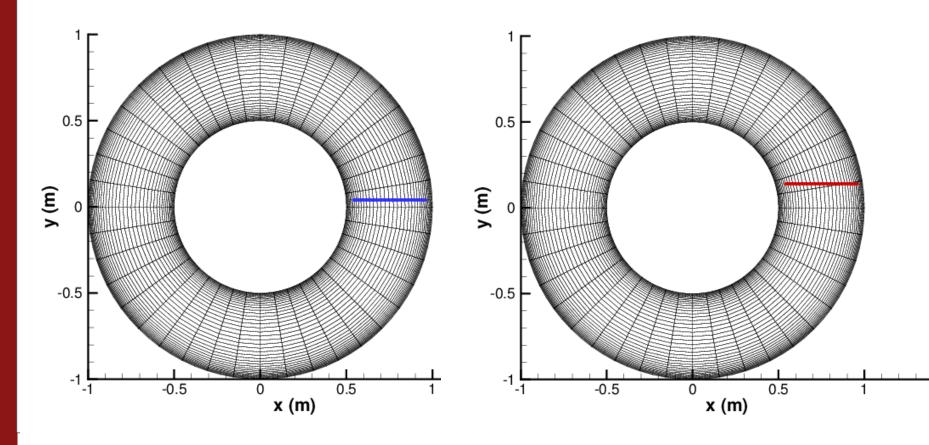
Compare straight-sided and curved elements



Test case: particle advection in annular channel

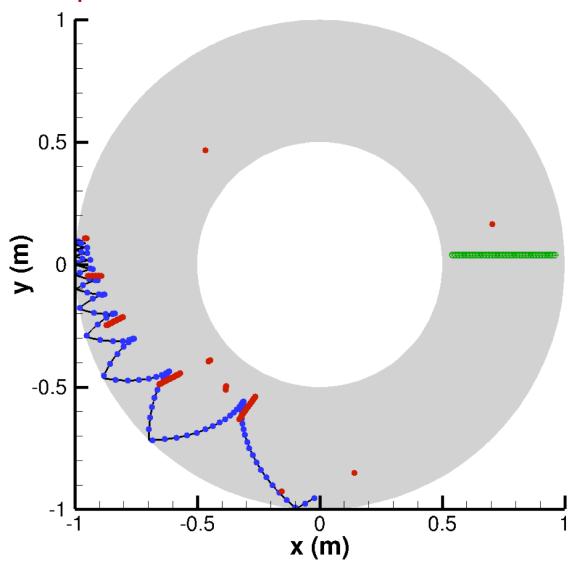
Blue: curved elements

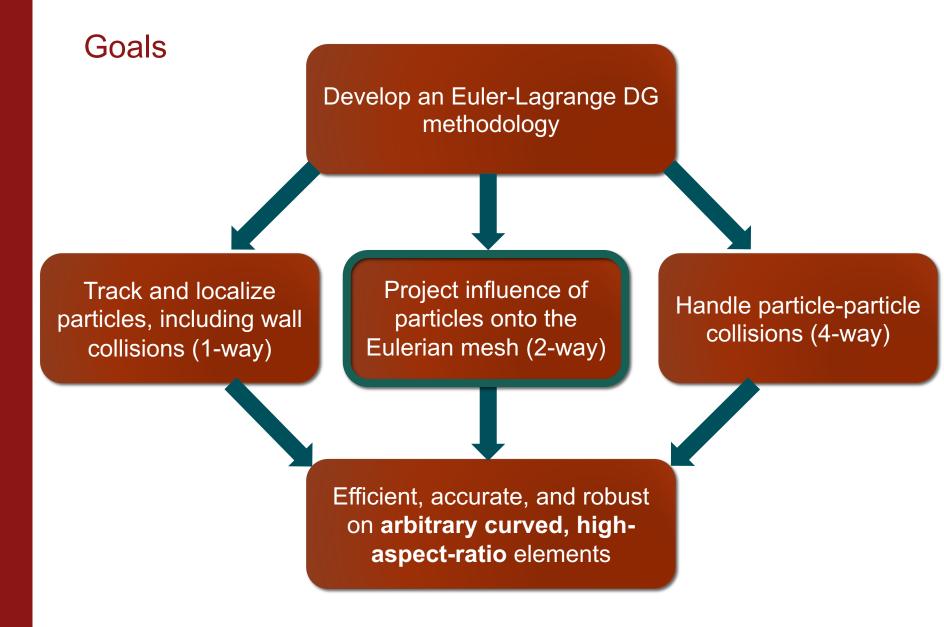
Red: straight-sided elements



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Test case: particle advection in annular channel





Reverse coupling

Project the influence of the particles onto the Eulerian mesh

Source term
$$\boldsymbol{S} = [0, \boldsymbol{S}_m, S_e]^T$$

Projection kernel $\chi(\boldsymbol{x}; \boldsymbol{x}_{d,i})$

$$lacksquare rg \max_{oldsymbol{x} \in \Omega} \chi(oldsymbol{x}; oldsymbol{x}_d) = oldsymbol{x}_d$$

$$oldsymbol{S}_m = -\sum_{i=1}^{N_p} oldsymbol{F}_i \chi(oldsymbol{x}; oldsymbol{x}_{d,i}),$$

$$S_E = -\sum_{i=1}^{N_p} igg(Q_i + oldsymbol{u}_{d,i} \cdot oldsymbol{F}_iigg) \chi(oldsymbol{x}; oldsymbol{x}_{d,i})$$

Finite-volume context

- Box kernel¹
- Volume-weighted², distance-weighted³ kernel
- Isotropic Gaussian function⁴

^[1] Crowe, J Fluids Eng, 1982

^[2] Squires and Eaton, Phys Fluid A, 1990

^[3] Elghobashi and Truesdell, Phys Fluid A, 1993

^[4] Capecelatro and Desjardins, J Comp Phys, 2013

Reverse coupling

DG context

- Quadrature used to evaluate integrals
- Multiple DOFs per element, subcell resolution
- Modal basis functions
- More susceptible to numerical noise and instabilities
- Isotropic Gaussian function^{1,2,3}

Shifted delta function
$$\chi(\boldsymbol{x}; \boldsymbol{x}_{d,i}) = \delta(\boldsymbol{x} - \boldsymbol{x}_{d,i})$$

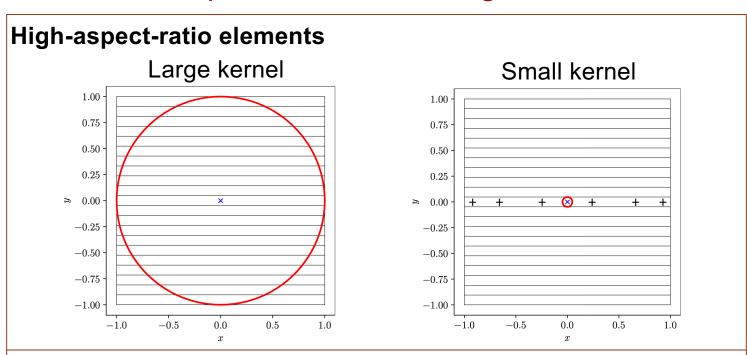
$$\int_{\Omega_e} \phi_m \boldsymbol{S} d\Omega_e = \sum_{i=1}^{N_p^e} \phi_m(\boldsymbol{x}_{d,i}) [0, \boldsymbol{F}_i, Q_i + \boldsymbol{u}_{d,i} \cdot \boldsymbol{F}_i]^T$$

- Simple
- Not smooth → numerical noise
- Inappropriate for larger particles

^[2] Huang et al, Comput Methods Appl Mech Eng, 2019

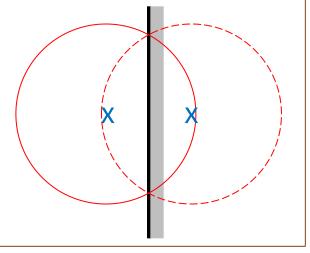
^[3] Zwick and Balachandar, Int J High Perform Comput Appl, 2020

Smooth isotropic kernels: challenges



Curved elements

- To conserve near-wall interphase transfer, common to use mirror particles on straight-sided elements
- Not straightforward for arbitrary curved elements

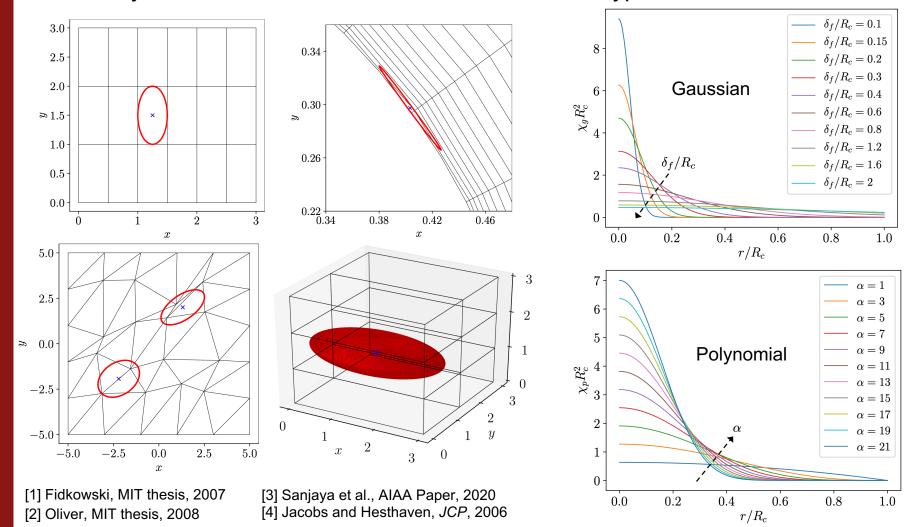


Proposed methodology

- 1. Smooth anisotropic kernels
 - Balance:
 - Accuracy
 - Mitigation of numerical noise and instabilities
 - Computational cost
 - Elliptical/ellipsoidal (instead of circular/spherical) in 2D/3D
- 2. Efficient kernel rescaling near walls
 - Avoid brute-force approach
 - Employ high-order polynomials to approximate rescaling factor

Smooth anisotropic kernel

- Construct ellipse based on mesh-implied metric^{1,2,3}
- Polynomial-based kernel⁴ instead of Gaussian-type kernel



Near-wall treatment

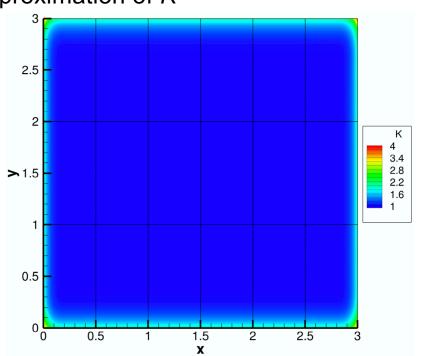
Brute-force approach: exact rescaling

$$K(oldsymbol{x}_d) = rac{1}{\sum\limits_{e=1}^{N_v} \int_{\Omega_e} \chi_p(oldsymbol{x}; oldsymbol{x}_d) d\Omega}$$

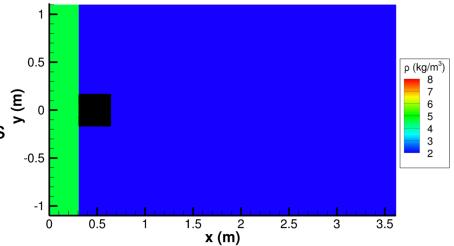
Proposed approach: approximate rescaling

- Construct high-order polynomial approximation of K
 - Sample K at quadrature points
 - Project to high-order polynomials

$$K_h^e(oldsymbol{x}_d) = \sum_{n=1}^{N_K} \widetilde{K}_n^e \Phi_n(oldsymbol{x}_d)$$

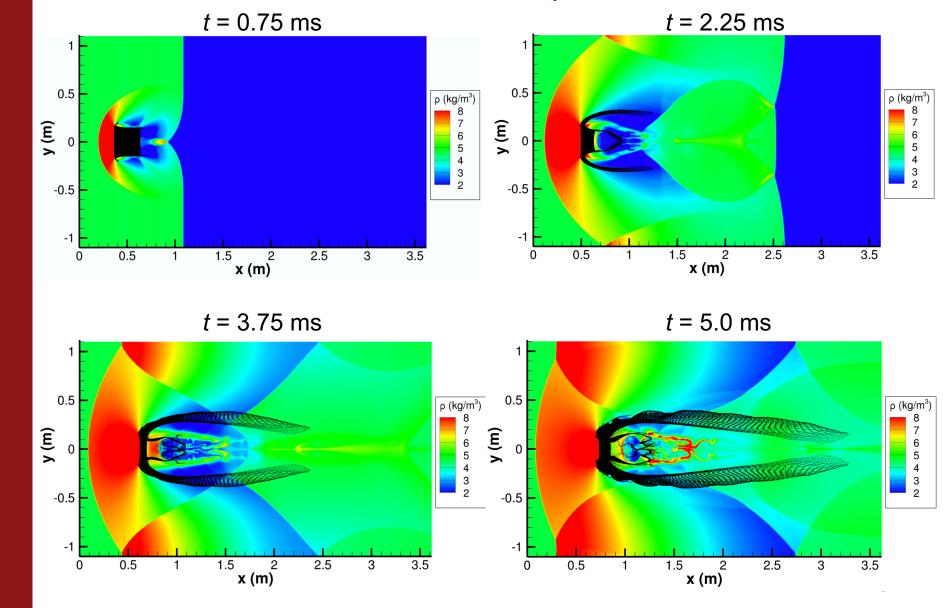


- Bronze particle cloud
- Volume fraction = 4%
- 500,000 2D quadrilateral elements
- p = 3 polynomials

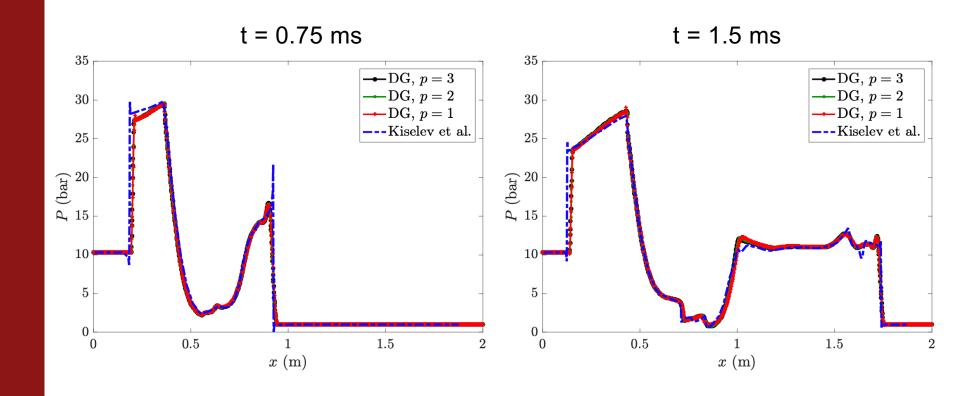


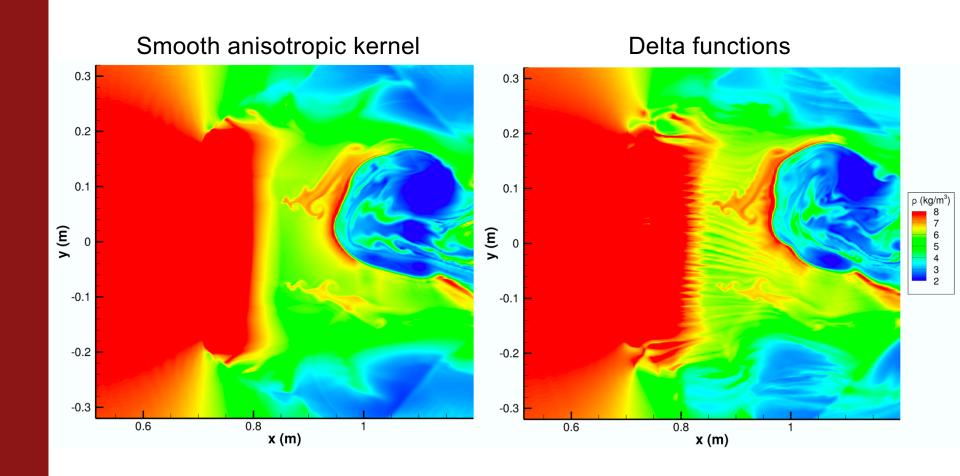
$\rho_d \; (\mathrm{kg/m^3})$	$D (\mu m)$	$c_d (\mathrm{J/kg/K})$	$\rho_c \; (\mathrm{kg/m^3})$	$u_c \text{ (m/s)}$	P_c (bar)
8900	100	435	1.2	0	1

- Difference between delta shape function and smooth anisotropic kernel
- Good agreement with numerical results¹

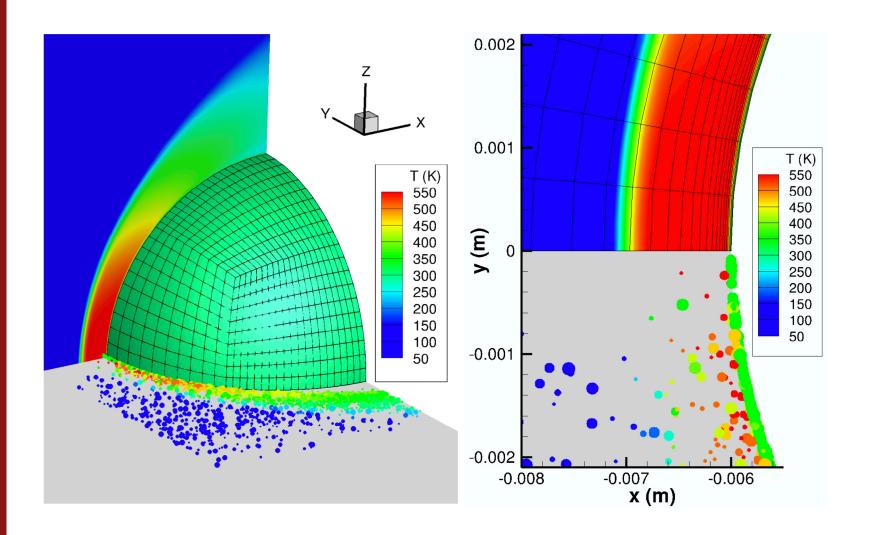


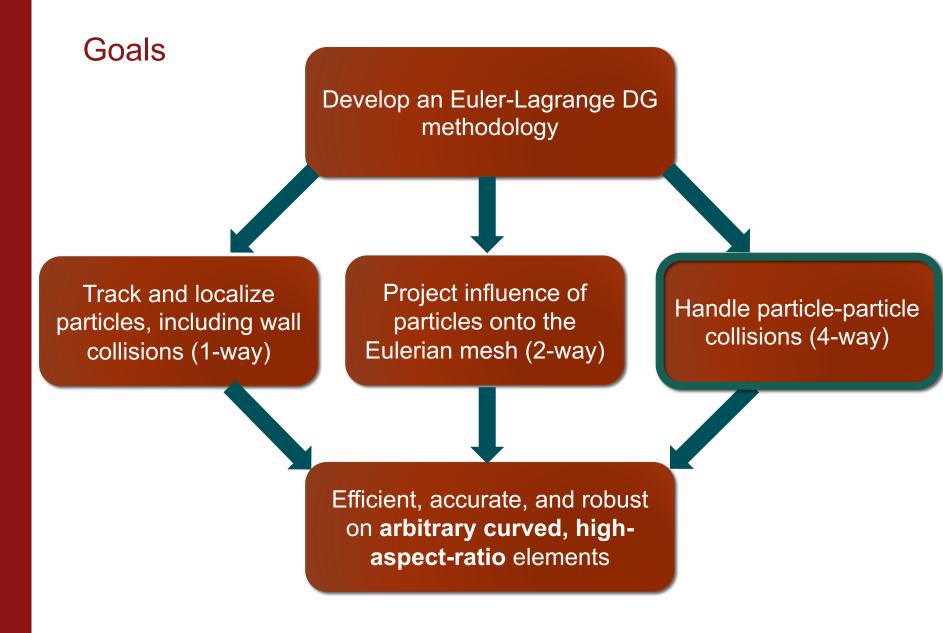
Comparison with numerical results





Hypersonic dusty flows over blunt bodies





Particle-particle collisions

Hard-sphere model

Select which particle pairs to inspect

Inspect particle pairs for potential collisions during given time step

Update collisions involving the recently collided particles

Sort potential collisions in chronological order

Enact first collision

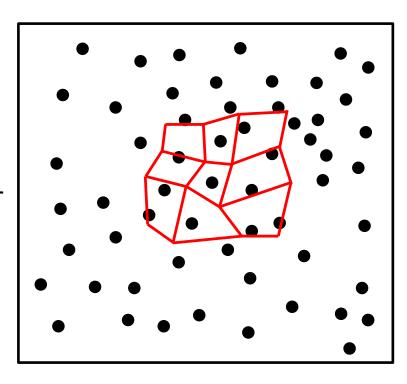
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Selecting which particle pairs to inspect

- Brute-force approach
 - All possible particle pairs in domain
- Element neighbor lists
 - Inspect particles in node-sharing elements

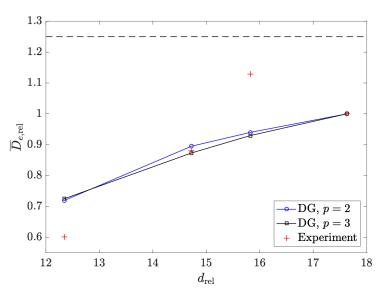
Truncate element neighbor lists

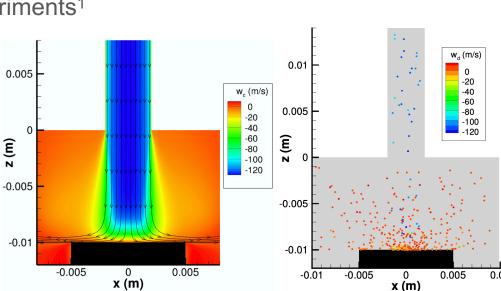
- Exploit the geometric mapping to further localize pairings
- Compatible with unstructured curved elements
- Simple and fast truncation process



Test cases

- 2D/3D kinetic theory example
 - Recover equilibrium properties
 - Significant speedup
- 3D sandblasting example
 - Impingement of particle stream on a flat plate
 - Effect of particle mass flux on erosion
 - Good agreement with experiments¹





0.2

0.3

0.4

 $|\boldsymbol{u}_d| \; (\mathrm{m/s})$

0.1

3.5

 $(m/s)^{2.5}$ $_{2}^{2}$ $_{1.5}^{2}$

0.5

Stanford University

Simulation

Kinetic theory

0.7

0.8

Summary

Developed an Euler-Lagrange DG solver

- Simple and robust shock capturing
 - Accurate surface heating predictions in hypersonic flows
- Lagrange particle methods for 1-way, 2-way, 4-way coupling
 - Compatible with arbitrary curved, high-aspect-ratio elements

Why DG?

- DG can considerably simplify meshing of complex geometries
- Curved elements can significantly improve predictions of particle trajectories

Outline

Development Background **Application** Hypersonic dusty flows over blunt bodies Introduction & Shock capturing Motivation scheme Experimental data and parametric Lagrangian **DG** mathematical sensitivities particle method background Mars atmospheric entry

Hypersonic dusty flows over blunt bodies

Relevant to Mars atmospheric entry

Adverse effects¹:

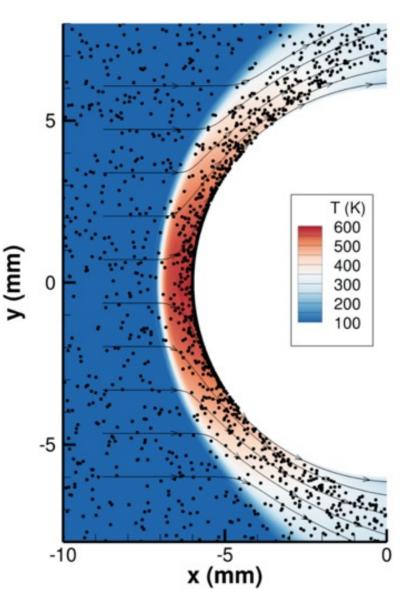
- Surface erosion
 - Particle-wall collisions
- Surface heat flux augmentation
 - Particle-wall collisions
 - Two-way coupling

Limited high-quality experimental data

 Ambiguity in how to reliably model these flows

Goals:

- Simulate the hypersonic dusty flow experiment by Vasilevskii et al.²
- Evaluate solution sensitivities to physical modeling of the particle phase



[1] Montois et al., International Planetary Probe Workshop, 2007 [2] Vasilevskii et al., *J Eng Phys Thermophys*, 2001.

Experiments by Vasilevskii et al.

Hypersonic dusty flow over a sphere

Gas	Ma_{∞}	$P_{t,\infty}$ (bar)	$T_{t,\infty}$ (K)	R_s (m)
N_2	6.1	17.5	570	0.006

Dust material	$\rho_d \; (\mathrm{kg/m^3})$	\overline{D} (m)	β
SiO_2	2264	0.19×10^{-6}	0.03

Measured ratio of dusty-gas to pure-gas heat flux at stagnation point

Solver details

- p = 2
- 60,000 elements
- Implicit-explicit time stepping
 - BDF3-AB3
- Two-way coupling

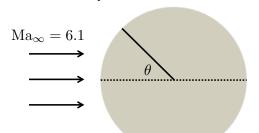
Baseline physical model

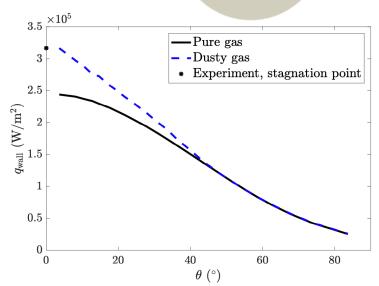
Simplest model that gives good agreement with experiment

- Henderson drag correlation¹
- Fox Nusselt number correlation²
- Thermophoretic force³

$$m_d \frac{d \boldsymbol{u}_d}{dt} = \boldsymbol{F} = \boldsymbol{F}_{qs} + \boldsymbol{F}_{th}$$

$$m_d c_d \frac{dT_d}{dt} = Q = Q_{qs}$$





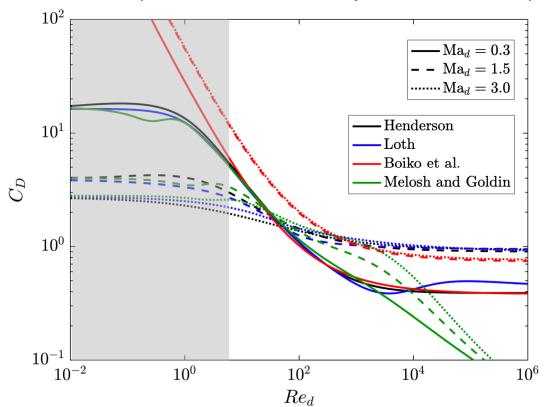
Sensitivity study to be conducted with respect to this baseline model

^[1] Henderson, AIAA J, 1976

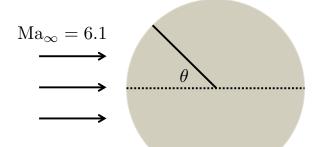
^[2] Fox et al., 11th International Shock Tubes and Waves Symposium, 1978

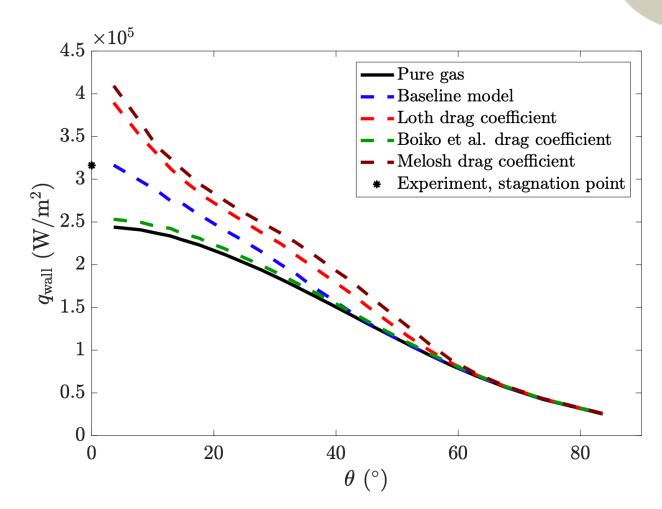
Drag correlations

- Henderson (AIAA J, 1976)
- Loth (AIAA J, 2008)
- Boiko et al. (Shock Waves, 1997)
- Melosh and Goldin (Lunar and Planetary Science, 2008)



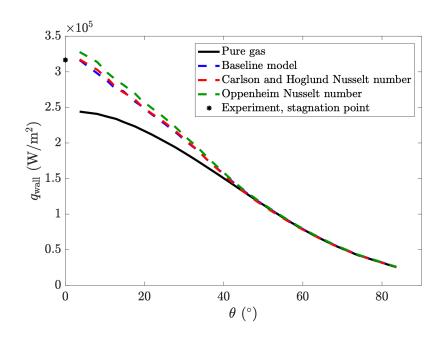
Sensitivity to drag correlation

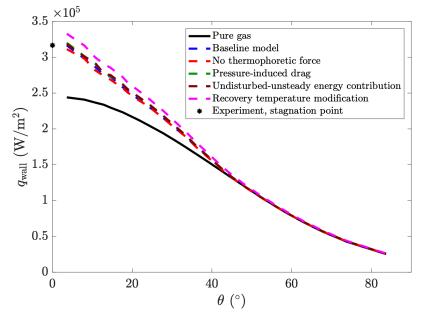




Additional sensitivities

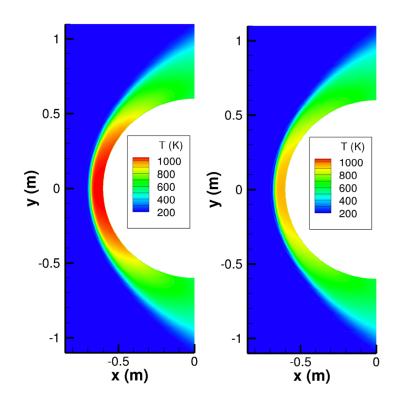
- Nusselt number correlations
 - Small sensitivity
- Momentum and energy contributions
 - Quasi-steady drag and heating most important





Additional sensitivities

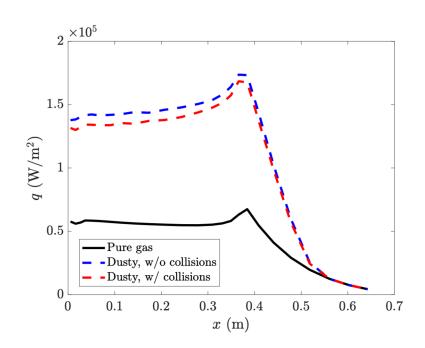
- Nusselt number correlations
 - Small sensitivity
- Momentum and energy contributions
 - Quasi-steady drag and heating most important
- Gas model
 - Shock standoff distance and shock layer quantities can have significant influence

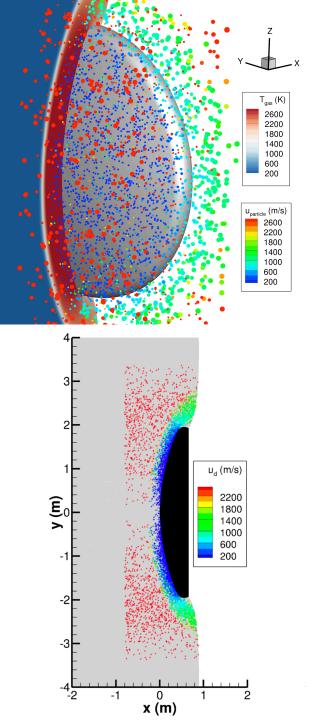


Additional sensitivities

Particle-particle collisions

- Hypersonic dusty flow over a capsule forebody
- Can attenuate particle-wall collisional energy flux
- Only important at very high dust loading





Outline

Development Background **Application** Hypersonic dusty flows over blunt bodies Introduction & Shock capturing Motivation scheme Experimental data and parametric Lagrangian **DG** mathematical sensitivities particle method background Mars atmospheric entry

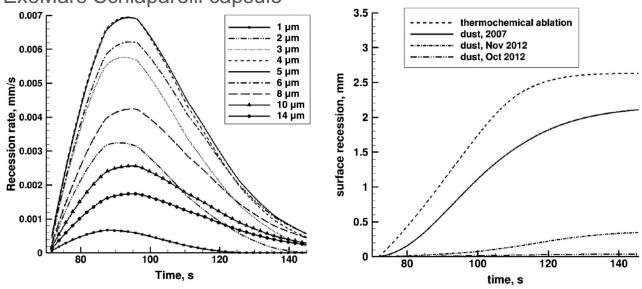
Background

Build on:

Goals:

- Previous section
- Palmer et al., Journal of Spacecraft and Rockets, 2020

 Computed heat shield recession at the stagnation point during the trajectory of the ExoMars Schiaparelli capsule



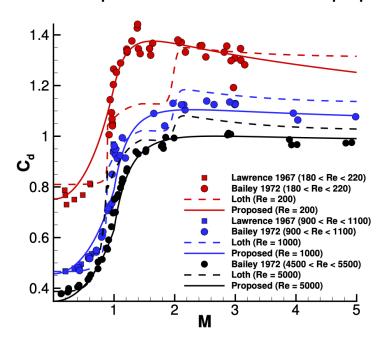
- Further improve understanding of relevant physical processes
- Assess effects of particle size distribution, angle of attack, and two-way coupling on heating augmentation and erosion
- Employ recently developed physics-based drag correlation¹
- Investigate particle trajectories in aft region

Drag correlation

Recently developed physics-based drag correlation¹ that incorporates:

- Low-speed hydrodynamics
- Shockwave physics
- Rarefied gas dynamics

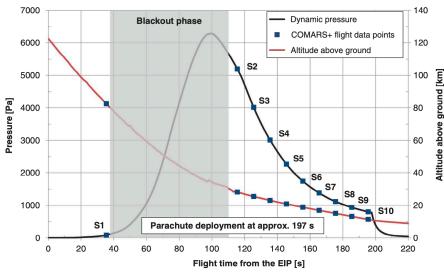
Better agreement with experimental data than popular correlations^{2,3}



ExoMars Schiaparelli trajectory

- 1st mission of the European Space Agency's ExoMars program
- Goals:
 - Search for evidence of methane and other gases that could signify biological or geological processes
 - Test key technologies to support future missions





ExoMars Schiaparelli trajectory

S3 trajectory point¹

Gas	Mach	u (m/s)	P (Pa)	T (K)
CO_2	8.97	2014	74.1	195

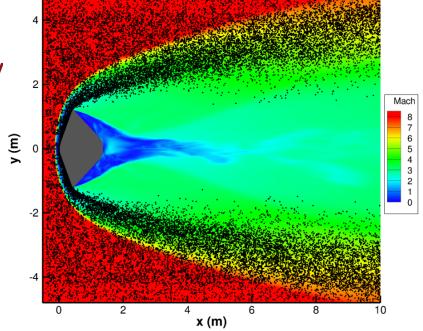
Two angles of attack: 0 and 20 deg

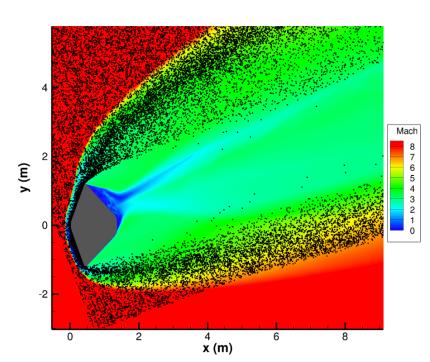
SiO₂ particles

Particle-wall impacts

- Heating augmentation by collisional energy transfer
- Surface recession²

~300,000 3D hexahedral elements Third-order-accurate (*p* = 2) polynomials

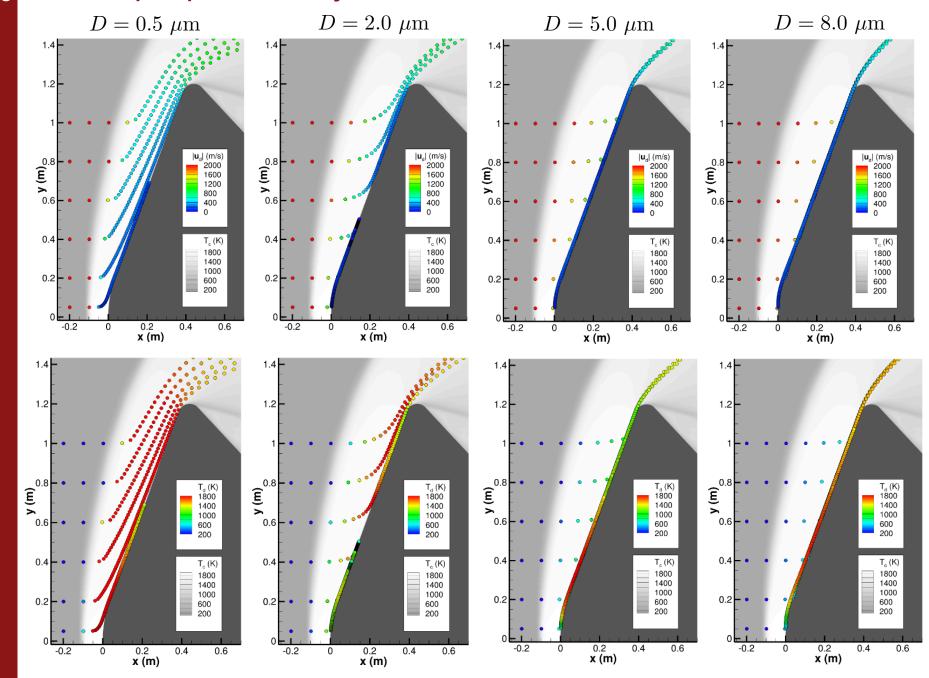




^[1] Gulhan et al., J Spacecraft Rockets, 2019

^[2] Palmer et al., J Spacecraft Rockets, 2020

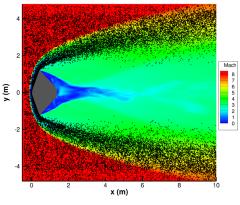
Sample particle trajectories



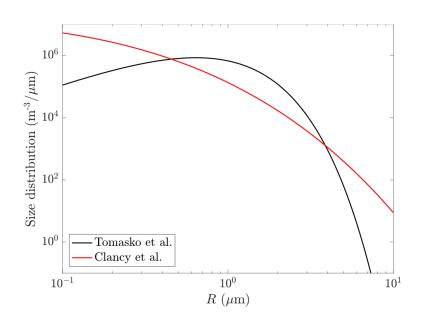
Effect of particle size distribution

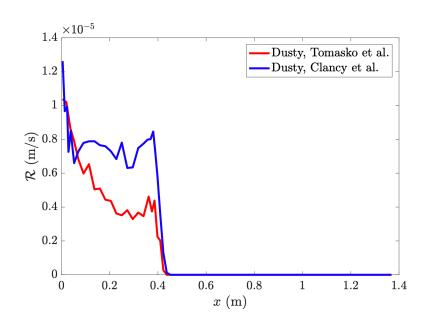
Modified gamma distribution:

- Tomasko et al., J Geophys Research: Planets, 1999
- Clancy et al., J Geophys Research: Planets, 1995



Mass loading ratio (particle mass flux to gas mass flux): $\beta = 0.01\%$



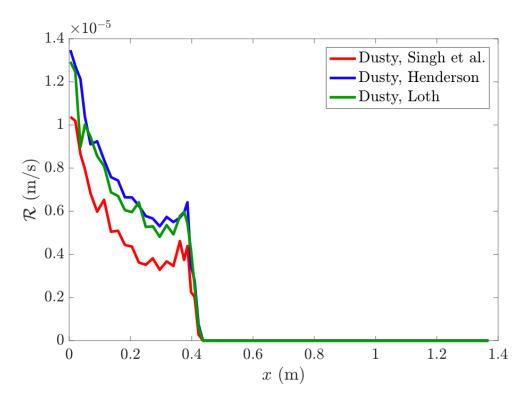


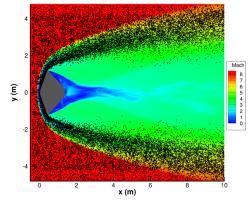
Stanford University

Effect of drag correlation

Drag correlations by:

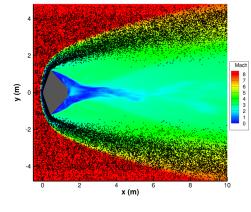
- Singh et al., https://arxiv.org/pdf/2012.04813
- Henderson, AIAA J, 1976
- Loth, *AIAA J*, 2008

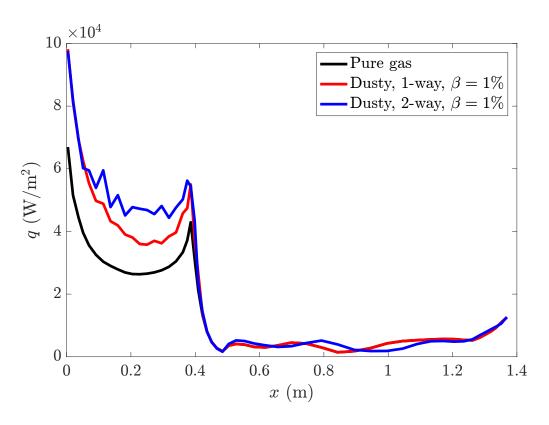




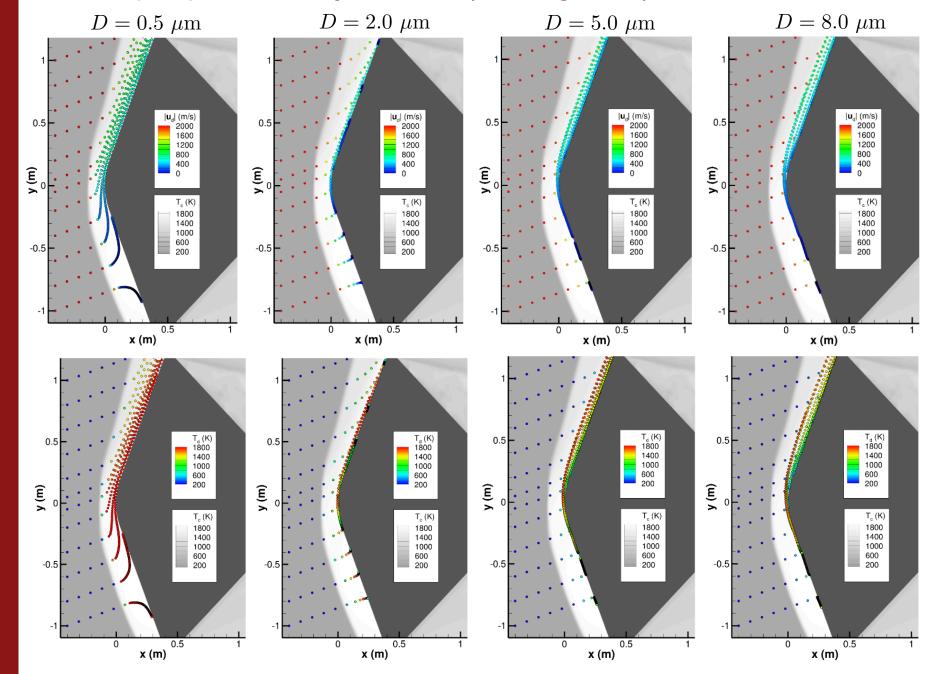
Effect of two-way coupling

Particles also affect carrier gas β = 1% mass loading ratio





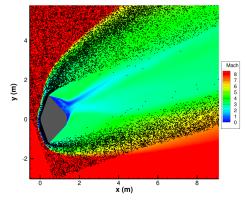
Sample particle trajectories (20 deg AoA)

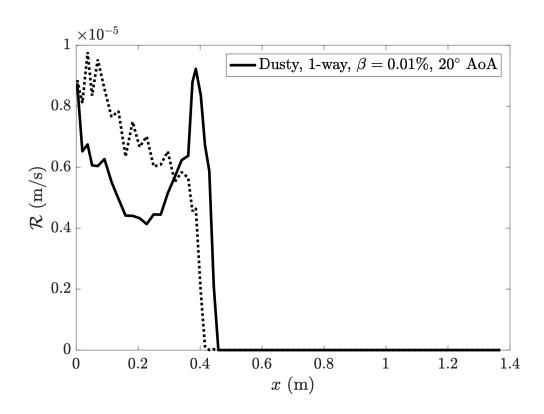


Surface recession

Solid line: windward side

Dotted line: leeward side





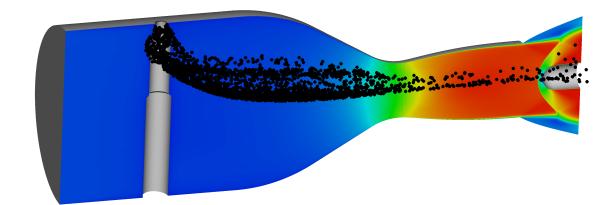
Main findings

- Dust impacts can cause surface erosion and heating augmentation
- Drag correlation by Singh et al. predicts less surface recession than popular correlations
- No direct interaction between particles and wake
- Large particles traverse the shock layer at higher speeds, lower temperatures than small particles
- Particle reverse-coupling at high dust loading can cause additional heating augmentation by transferring thermal energy to the boundary layer
- At nonzero angle of attack, leeward-side surface recession can be noticeably higher than windward-side recession

Summary & Outlook

Summary

- Developed a simple and robust shock capturing method for DG schemes
- Developed an Euler-Lagrange methodology in a DG framework suited for arbitrary curved, high-aspect-ratio elements
- Established significant benefits of the proposed DG methodology in the context of viscous hypersonic flows and particle-laden flows
- Applied the proposed methodology to hypersonic dusty flows over blunt bodies, with special application to Mars atmospheric entry



Outlook

- Further experimental validation: collaboration with DLR and NASA Ames
- Extend temperature range: thermochemical nonequilibrium

Thank you!

QUESTIONS

